

TriggerScope: Towards Detecting Logic Bombs in Android Applications

Yanick Fratantonio, Antonio Bianchi, William Robertson, Engin
Kirda, Christopher Kruegel, Giovanni Vigna

Pietro De Nicolao

Politecnico di Milano

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TriggerScope: Towards Detecting Logic Bombs in Android Applications

Yanick Fratantonio*, Antonio Bianchi*, William Robertson[†], Engin Kirda[†], Christopher Kruegel*, Giovanni Vigna*

*UC Santa Barbara

{yanick,antonio,chr,Chris,vigna}@cs.ucsb.edu

[†]Northeastern University

{wkr,ek}@ccs.neu.edu

Introduction to logic bombs

Logic bombs

Logic bomb: malicious application logic that is triggered only under certain (narrow) conditions.

- ▶ *Malicious application logic:* violation of user's reasonable expectations
- ▶ Malware is designed to target **specific victims, under certain circumstances**

Example: take a navigation application, supposed to help a soldier in a war zone find the shortest route to a location.

- ▶ after a given (hardcoded) date, it gives to him longer, more dangerous routes

Another (real) example

RemoteLock: Android app that allows the user to remotely lock and unlock the device with a SMS containing an user-defined keyword

- ▶ The app code also contains the following check:
`(!= (#sms/#body equals "adfbdfgfsghytdfsw")) 0)`
 - ▶ The predicate is triggered when an incoming SMS contains that hardcoded string
- ▶ By sending a SMS containing that string, the device unlocks
- ▶ That's a backdoor implemented with a **logic bomb**!

Problems with traditional defenses

App Stores employ some defenses, but they are not sufficient.

- ▶ **Static analysis:** malicious application logic doesn't require additional privileges or make "strange" API calls
 - ▶ Malicious behavior is deeply hidden within the app logic
 - ▶ Example: the malicious navigation app... behaved like any navigation app!
- ▶ **Dynamic analysis:** likely won't execute code triggered only on a future date or in a certain location
 - ▶ Code coverage problems
 - ▶ Can be detected and evaded
 - ▶ Even if covered, how to discern malicious behavior from benign?
- ▶ **Manual audit:** if source code is not available, no guarantees
 - ▶ Code can be obfuscated

TriggerScope

Key observation

TriggerScope detects logic bombs by precisely analyzing and characterizing **the checks (conditionals, predicates) that guard a given behavior**.

- ▶ It gives less importance to the (malicious) behavior itself.

```
if(sms.getBody().equals("adfbdf...")) // Look here!
{
    myObject.doSomething(); // ...not there.
}
```

Trigger analysis

- ▶ **Predicate:** logic formula used in a conditional statement
 - ▶ `(&& (!= (#sms/#body contains "MPS:") 0) (!= (#sms/#body contains "gps") 0))`
 - ▶ **Suspicious predicate:** a predicate satisfied only under very specific, narrow conditions
- ▶ **Functionality:** a set of basic blocks in a program
 - ▶ **Sensitive functionality:** a functionality performing, directly or indirectly a sensitive operation
 - ▶ In practice: all calls to Android APIs protected by permissions, and operations involving the filesystem
- ▶ **Trigger:** suspicious predicate controlling the execution of a sensitive functionality

Analysis overview (1)

1. **Static analysis** of bytecode; building of Control Flow Graph
2. **Symbolic Values Modeling** for integer, string, time, location and SMS-based objects
3. **Expression Trees** are built and appended to each symbolic object referenced in a check
 - ▶ Reconstruction of the *semantics* of the check, often lost in bytecode



Figure 1: Example of expression tree.

Analysis overview (2)

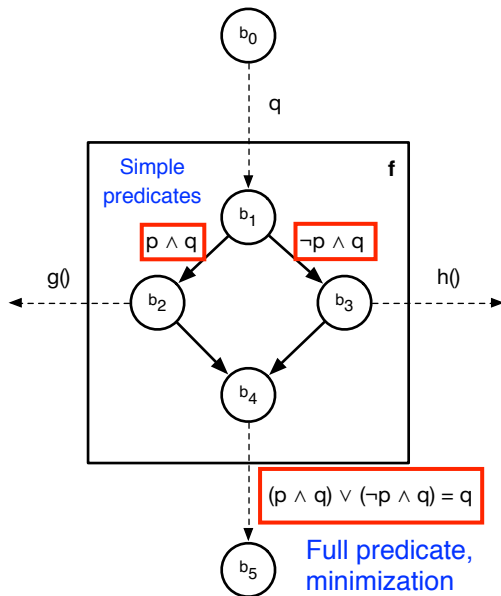
4. **Block Predicate Extraction:** edges of Control Flow Graph are annotated with simple predicates

- ▶ Simple predicate: P in `if P then X else Y`

5. **Path Predicate Recovery and Minimization**

- ▶ Combine simple predicates to get the *full path predicate* that reaches each basic block
- ▶ Minimization: elimination of redundant terms in predicates
 - ▶ important to reduce false dependencies

Path Predicate Recovery and Minimization



Analysis overview (3)

6. **Predicate Classification:** a check is **suspicious** if it's equivalent to:
 - ▶ Comparison between current *time* value and constant
 - ▶ Bounds check on GPS *location*
 - ▶ Hard-coded patterns on body or sender of *SMS*
7. **Control-Dependency Analysis:** control dependency between *suspicious predicates* and *sensitive functionalities*.
 - ▶ sensitive = privileged Android APIs + filesystem ops
 - ▶ **Suspiciousness propagates** with data flows and callbacks
 - ▶ Problem: data flows through files
 - ▶ When in doubt: suspicious!
8. **Post-processing:** whitelisting for some edge cases

Experiment

Data sets

- ▶ **Benign applications:** 9582 apps from Google Play Store
 - ▶ They all use time-, location- or SMS-related APIs
 - ▶ Actually, TriggerScope identified backdoors in two “benign” apps, confirmed by manual inspection!
- ▶ **Malicious applications:** 14 apps from several sources
 - ▶ Stealthy malware developed for previous researches
 - ▶ Real-world malware samples
 - ▶ HackingTeam RCSAndroid

Results of analysis

Analysis step	TP	FP	TN	FN	FPR	FNR
Predicate detection	14	1386	7927	0	14.88%	0%
Suspicious Predicate A.	14	462	8851	0	4.96%	0%
Control-Dependency A.	14	117	9196	0	1.26%	0%
TriggerScope (all)	14	35	9278	0	0.38%	0%

Table 1: Results of analysis after each step. Note how each step is useful to refine the analysis.

$$FPR = \frac{FP}{FP+TN}, FNR = \frac{FN}{FN+TP}$$

False Positives decreasing step after step

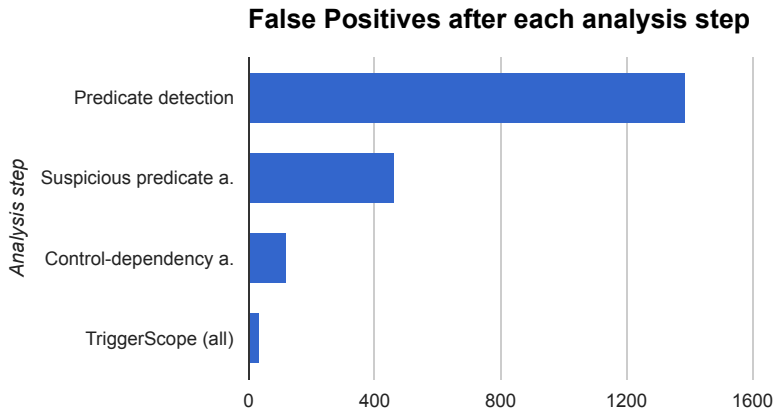


Figure 3: Each analysis stage is useful, because it reduces False Positives.

Critique

Strengths

- ▶ TriggerScope provides **rich semantics** on predicates
 - ▶ Great help for **manual audits**
 - ▶ This makes the tool extensible, open for future research
- ▶ Novel approach: **focus on checks**, not malicious behaviors
- ▶ **Fewer FPs, FNs** than other tools

Issues: limits of analysis

- ▶ Definition of **suspicious predicate** is too narrow
 - ▶ Only checks against hardcoded values are considered
 - ▶ Triggers can come from the network or elsewhere
- ▶ Authors claim **0% FNs**, but the evaluation isn't conclusive
 - ▶ *we manually inspected a random subset of 20 applications for which our analysis did not identify any suspicious check. We spent about 10 minutes per application, and we did not find any false negatives.*
 - ▶ Difficult to assess FNs if no tool finds anything and source code is unavailable
- ▶ This analysis is still **blacklisting**: listing things we don't like
 - ▶ We're competing against attackers' creativity

Issues: evasion techniques

- ▶ *Reflection, dynamic code loading, polymorphism and conditional code obfuscation* [Sharif, 2008] can defeat static analysis.
 - ▶ Authors say that these techniques are themselves suspicious, but they also have legitimate uses
- ▶ Predicate minimization is **NP-complete**
 - ▶ Is it possible to design “pathological” code to slow down and defeat analysis?
 - ▶ Or result in very complex, meaningless predicates?
- ▶ **Exceptions** were not cited as a control flow subversion method
 - ▶ *Statically reasoning in the presence of exceptions and about the effects of exceptions is challenging* [Liang, 2014]
 - ▶ Unclear how the static analysis engine handles exceptions
 - ▶ Unchecked exceptions (e.g. division by zero) could be exploited as stealthy triggers

Related and future work

Related work: AppContext

AppContext [Yang, 2015]: supervised machine learning method to classify malicious behavior statically

1. Starts identifying suspicious actions
2. Context: which category of input controls the execution of those actions?

Similar idea: just looking at the action isn't enough. Differences:

- ▶ AppContext only **classifies** triggers as suspicious or not; TriggerScope also provides **semantics** about the predicates, helping manual inspection
- ▶ AppContext does not consider the typology of the predicate, only the type of its inputs
- ▶ **Higher FP rate** than TriggerScope

Future evolutions

- ▶ **Extend trigger analysis** not only to time, location, SMSs
 - ▶ The trigger could come e.g. from the network
 - ▶ The framework is easily extensible to other types of triggers with more work to model other symbolic values
 - ▶ Is there **a more general approach**?
- ▶ **Quantitative analysis** of predicate “suspiciousness”
 - ▶ Currently, it’s defined in a qualitative, ad-hoc way
 - ▶ Could be combined with classification methods

References



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